

**Regional Precipitation Trends:
Distinguishing Natural Variability from Anthropogenic Forcing**

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ABSTRACT

In this study, the nature and causes for observed regional precipitation trends during 1977-2006 are diagnosed. It is found that major features of regional trends in annual precipitation during 1977-2006 are consistent with an atmospheric response to observed sea surface temperature (SST) variability. This includes drying over the eastern Pacific Ocean that extends into western portions of the Americas related to a cooling of eastern Pacific SSTs, and broad increases in rainfall over the tropical eastern hemisphere including a Sahelian rainfall recovery and increased wetness over the Indo-West Pacific related to North Atlantic and Indo-west Pacific ocean warming. It is further determined that these relationships between SST and rainfall change are generally not symptomatic of human-induced emissions of greenhouse gases (GHGs) and aerosols. The intensity of regional trends simulated in climate models using observed time variability in greenhouse gases, tropospheric sulphate aerosol, solar and volcanic aerosol forcing are appreciably weaker than observed and also weaker than simulated in atmospheric models using only observed SST forcing. The pattern of rainfall trends occurring in response to such external radiative forcing also departs significantly from observations, especially a simulated increase in rainfall over the tropical Pacific and southeastern Australia that are opposite in sign to the actual drying in these areas.

Additional experiments illustrate that the discrepancy between observed and GHG-forced rainfall changes during 1977-2006 results mostly from the differences between observed and externally-forced SST trends. Only weak rainfall sensitivity is found to occur in response to the uniform distribution of SST warming that is induced by GHG and aerosol forcing, whereas the particular pattern of observed SST change that includes increased SST contrast between the east Pacific

and the Indian Ocean, and strong regional warming of the North Atlantic Ocean were key drivers of rainfall regional trends. The results of this attribution study on the causes for 1977-2006 regional rainfall changes are used to discuss prediction challenges including the likelihood that recent rainfall trends might persist.

1. Introduction

Regional precipitation is sensitive to sea surface temperatures (SSTs) on the interannual to multi-decadal time scale. El Niño/ Southern Oscillation (ENSO), for instance, is the main attributable cause of tropical droughts and floods that occur interannually (e.g. Ropelewski and Halpert 1986, 1987 among many studies). A characteristic east-west dipole of Indian Ocean SST anomalies, another natural mode of coupled ocean-atmosphere variability occurring interannually, is believed to be a factor causing southeast Australian drought events (Ummenhofer et al. 2009). Decadal-long droughts and pluvials over southwestern North America and the Great Plains have also been attributed to tropical Pacific and North Atlantic SST conditions (Schubert et al. 2004a, 2004b; Seager et al. 2005), though the decade-long severity of the “Big Dry” over a large portion of Australia since 1995 appears unexplainable by Indian Ocean SST impacts alone (Ummenhofer et al. 2009). On multidecadal time scales, there is substantial evidence that much of the Sahelian drying trend during the latter half of the 20th Century resulted from slow variations in North Atlantic and Indian Ocean SSTs (e.g., Giannini et al. 2003; Hoerling et al. 2006). And key regional and seasonal features of U.S. precipitation trends during 1950-2000 have also been linked to multidecadal variability in Pacific and Atlantic SSTs (Wang et al. 2009).

Explaining the nature of regional precipitation change is key for decision makers who seek climate information to help guide their adaptation and mitigation strategies. Of particular importance is ensuring that natural variability, when occurring, is not misunderstood to indicate that climate change is either not happening or is happening more intensely than the true human influence. Efforts to detect human influences on regional precipitation change on centennial time scales have so far been unsuccessful mainly due to a weak signal-to-noise ratio (Hegerl et

al. 2007). There is some indication for the anthropogenic forcing of 20th Century changes in precipitation averaged within latitude bands (Zhang et al. 2007), perhaps related to a widening of the tropical belt (Seidel and Randel 2007; Seidel et al. 2007). However, the amplitude of 20th Century trends in zonally averaged precipitation are much larger than can be explained by an anthropogenic influence alone. An assessment by the U.S. Climate Change Science Program of the causes of North American precipitation trends during the last half-century concluded that spatial variations and seasonal differences in change were unlikely the result of anthropogenic forcing alone, but that some of the variations likely resulted from regional SST forcing (CCSP 2009).

In this paper we diagnose the factors contributing to regional precipitation trends since the mid-1970s, a period of anthropogenic warming of the climate system as a whole that includes a substantial increase in global SSTs (IPCC 2007). We inquire whether regional precipitation trends have been consistent with SST changes, and where such associations exist, assess if they are symptomatic of natural variability and/or anthropogenic forcing. Figure 1 shows the 30-yr trend in observed annual SSTs (top) during 1977-2006 revealing a warming over most ocean basins. Conspicuous has been an absence of warming along the equatorial Pacific from the dateline to South America and an expanse of the east Pacific between 30°N and 30°S. By contrast, a relatively uniform pattern of SST warming is simulated in coupled models using observed and projected greenhouse gas (GHG) and aerosol concentrations and other forcings up through 2006 (Fig. 1, middle), whose spatial correlation with observations over the world oceans is only 0.32. The diagnosis is of the Coupled Model Intercomparison Project's (CMIP3; Meehl et al. 2007) archive of model data (which includes 22 coupled atmosphere-ocean models) using

projected climate forcing changes after 1999 based on a business-as-usual scenario (A1B) that extends from the historical climate of the 20th Century runs. Our interest is not to conduct a detailed attribution of the SST changes per se (see Knutson et al. 2006; Santer et al. 2006), but rather to assess the practical implications for the differences between observed and externally forced SST changes (Fig. 1, bottom) on regional precipitation changes.

The outline of the paper and main results are as follows. Section 2 describes the data sets and methods used to diagnose causes for observed regional precipitation trends during 1977-2006. Analysis of AMIP simulations in Section 3 illustrates that several major features of the observed annual precipitation trends are consistent with a SST forced signal. These include drying of the tropical east Pacific and adjacent portions of western North and South America, a recovery in Sahel rainfall, increased rainfall over the tropical Indo-west Pacific, and decreased precipitation over southeast Australia (though the latter SST signal is much weaker than the observed decline). Analysis of CMIP simulations reveals major features of observed annual precipitation trends to be mostly inconsistent with external radiative forcing. The CMIP ensemble mean yields increased rainfall over the equatorial east Pacific and southeast Australia during 1977-2006, contrary to the declines observed, whilst increased rainfall simulated over the Sahel and the tropical Indo-Pacific is an order of magnitude weaker than the observed increases. Section 3 also places the intensity of observed precipitation trends during the last 3 decades into the context of those resulting from unforced coupled ocean-atmosphere variability as well as from unforced atmosphere variability alone. Section 4 explores reasons for differences between AMIP and CMIP rainfall trends using additional climate simulations, the diagnosis of which reveals large sensitivity to the particular pattern of SST trends and establishes that the SST trend differences

seen in Fig. 1 are the main cause for their simulated rainfall trend differences. Concluding remarks address the nature of the recent SST trends, and we pose several questions germane to predicting regional precipitation change over coming decades.

2. Data and Methods

2.1 Observational data and analysis period

Global precipitation analysis is based on a merger of two data sets. The satellite precipitation product of the Global Precipitation Climatology Project (GPCP; Huffman et al. 1997) is used for ocean regions. These data are monthly from January 1979-present, and are gridded on 2.5° latitude x 2.5° longitude boxes. A particular strength of the GPCP product is the use of passive microwave satellite measurements which offer a more direct precipitation estimate compared to visible and infrared satellite measurements alone (WCRP 2008). The GPCP is also believed to be more realistic than the CMAP satellite product which has an artificial decreasing trend in global mean ocean precipitation (Yin et al. 2004), although one must bear in mind that existing satellite products were not specifically designed for trend analysis. The Global Precipitation Climatology Centre analysis (GPCC; Rudolf and Schneider 2005) is used for land regions. These data are monthly from January 1901-present, and are gridded at 0.5° resolution. Global monthly SST data is based on the UK Meteorological Office's HadISST2 1° gridded analysis (Rayner et al. 2003). We also consult two other SST analyses; the 5° gridded Kaplan product (Kaplan et al. 2000), and the 2.5° gridded NOAA product (Smith and Reynolds 2005).

There are practical issues involved in choosing the recent 1977-2006 period to assess rainfall trends. Foremost is that satellite data is required to provide a global view of precipitation, and

that source only becomes available after 1974 (based on NOAA polar orbiting sensors alone). As indicated above, the GPCP multi-sensor based dataset (1979-present) provides the best current satellite rainfall product for climate studies. The second issue concerns the model simulations. As we conceived this study of regional trends in climate, we wished to focus on the most recent 30-yr period (which then was 1977-2006). Both CCSP and IPCC AR4 reports have indicated that most of the increase in global and continental-scale land surface temperature and SST (e.g., the tropical warm pool) has occurred post-1970. It was thus deemed reasonable to focus an attribution effort on this most recent 30-year period that has experienced a detectable anthropogenic warming. It is certainly true that the selection of start and end years will influence trend estimates, especially for a short period as used herein. In this paper, we use the phrase “trend” in its standard sense, to indicate a general movement or tendency in the system over a period of time (in our case, 1977-2006), without necessarily suggesting that the observed trends represent a detected change in a *statistical* sense. We have taken the approach to determine whether the observed trends can be understood in a *physical* sense, related to either natural and/or anthropogenic influences, especially regarding the role of oceans.

2.2 *Climate model simulations*

Two configurations of climate model simulations are used to determine the causes for observed precipitation trends; atmospheric general circulation models (AMIP), and coupled ocean-atmosphere general circulation models (CMIP). For the former, a total of 4 different models were available, each subjected to specified monthly varying observed global SSTs, but climatological values for the chemical composition of the atmosphere. An equal sized 9-member ensemble was used for each model yielding a total ensemble of 36 runs during the 1977-2006

period of analysis¹. For the CMIP simulations, a total of 21 different models were utilized, each subjected to specified monthly variations in greenhouse gases, tropospheric sulphate aerosols, solar irradiance and the radiative effects of volcanic activity for 1880-1999, and with the IPCC Special Emissions Scenario (SRES) A1B (IPCC, 2007) thereafter². Only a few of the modeling centers generated multiple runs. To ensure equal weighting of all models, our analysis uses a single run from each of the modeling centers yielding a total ensemble of 21 members, with the model data accessed from the Program for Climate Model Diagnosis and Intercomparison (PCMDI) archive as part of the Coupled Model Intercomparison Project (CMIP3; Meehl et al. 2007). The externally forced (greenhouse gas, aerosol, solar and volcanic) signal in precipitation is estimated by averaging the multi-model CMIP ensemble members, whereas the SST-forced signal is estimated by averaging the multi-model AMIP ensemble members. The linear trend in precipitation is calculated for each model simulation, and subsequently averaged across all ensemble members.

Unforced control integrations of CMIP3 coupled models are also diagnosed. Most modeling centers generated roughly 300-yr simulations using climatological well-mixed greenhouse gases

¹ The 4 atmospheric models used are the NCAR Community Climate Model (CCM3; Kiehl et al. 1996), the NASA Seasonal-to-Interannual Prediction Project (NSIPP) model (Schubert et al. 2004a), the European Center/Hamburg model (ECHAM4.5, Roeckner et al. 1993), and the Experimental Climate Prediction Center's (ECPC) model (Kanamitsu et al. 2002).

² We did not include the MIUB-ECHO model because its upper troposphere circulation data were unavailable. Also, only half of the CMIP models included solar and volcanic forcing.

associated with pre-1900 conditions. We calculate the statistics of 30-yr precipitation trends from these experiments in order to estimate the influence of internal coupled ocean-atmosphere noise. Since ENSO is an important source for global precipitation variability, we select four different models that have been shown to possess realistic interannual SST variance in the tropical Pacific and also to exhibit realistic impacts on global precipitation variability (AchutaRao and Sperber 2006): the Canadian Center for Climate Modeling and Analysis CGCM3.1, the Meteorological Research Institute MRI-CGCM2.3, the National Center for Atmospheric Research PCM, and the Hadley Center for Climate Prediction and research/Met Office UKMO-HadCM3.

An additional suite of atmospheric climate model simulations are conducted that specify the 30-year change in SSTs for 1977-2006 as the sole anomalous boundary condition, with twin experiments performed that use either the observed or the CMIP SST change (see Fig. 1). We again employ four different atmospheric models, and for each generate a 30-member ensemble. The atmospheric sensitivity to the specified SST changes in these runs is determined by comparing with 50-year control integrations of each model that used a 1971-2000 climatologically averaged, seasonally varying SST.

3. Observed and Simulated Regional Precipitation Trends 1977-2006

a. Spatial patterns of 30-yr change

Figure 2 compares the observed (top), AMIP simulated (middle), and CMIP simulated (bottom) trends in annual precipitation. Table 1 summarizes observed and simulated mean regional changes together with estimates for the noise of 30-yr rainfall trends. Owing to the limited availability of satellite data, the observed precipitation trends over oceans are calculated for

1979-2006, whereas the entire 30-year period beginning in 1977 is used over land. The model trends have been calculated for the entire 30-yr period over both land and oceans, and the AMIP and CMIP ensemble trends over the oceans were confirmed not to be materially different when derived for the shorter post-1979 period.

The pattern of observed precipitation change over the tropical Pacific and adjacent portions of the Americas resembles the interannual pattern associated with La Niña (Ropelewski and Halpert, 1986, 1987). A narrow band of equatorial drying is consistent with the slight cooling of the cold tongue SSTs (cf. Fig. 1). Drying fans poleward over the subtropical North and South Pacific and extends over the adjacent land areas including the southwest United States and southern Brazil. In contrast, much of the tropical eastern hemisphere experienced an increase in annual rainfall. This includes the Sahel which saw some recovery from the severe drought years of the 1970s and '80s (Lebel and Ali 2008), and much of the Indo-west Pacific region. A drying trend over eastern Australia has been a noteworthy exception to this overall regional wettening.

Diagnosis of AMIP simulations reveals that several regional features of the observed rainfall trends are consistent in sign with the atmosphere's response to SST forcing (Fig. 2 middle). The ensemble mean trend includes reduced rainfall over the equatorial east Pacific with drying extending into the subtropics and the western portions of the Americas. Over the eastern hemisphere, Sahel rainfall increases in response to SST variability during 1977-2006, as does rainfall over the oceanic regions of the Indo-Pacific warm pool. Only a weak precipitation response is simulated over Australia suggesting that the observed changes over that continent during the recent 30-years are not consistent with SST forcing alone.

Considerably less sign agreement exists between the regional patterns of observed precipitation trends and those simulated in response to external radiative forcing (Fig. 2 bottom). We note the CMIP ensemble response to GHG forcing consists of increased rainfall over the equatorial Pacific, contrary to the drying observed and also contrary to the drying in AMIP simulations. Likewise, a simulated increase in east Australian rainfall is opposite to the observed drying. There is better sign agreement with the observed trends, and also with the AMIP simulated trends, over the Sahel and Indo-west Pacific warm pool (see Table 1). However, regarding the amplitude of the ensemble rainfall trends, the CMIP response is much weaker than observed, and is likewise weaker than in the AMIP ensemble simulations. The small amplitude compared to observations is not surprising given prior indications for a weak anthropogenic signal-to-noise ratio for zonally averaged precipitation trends (Zhang et al. 2006).

There are several factors that may explain the different responses between AMIP and CMIP experiments, including the curious result of a much weaker CMIP rainfall change over the Indo-west Pacific compared to AMIP. The general view is that tropical rainfall is sensitive to changes in SST *gradients* (for instance the large rainfall anomalies in association with the ENSO cycle during which appreciable changes in zonal SST gradients occur). We conjecture that the intensity of AMIP rainfall responses over the warm pool depends not only on the local increase of the Indo-Pacific SSTs during 1977-2006 (where CMIP SST change agrees well with observations), but also on the juxtaposed cooling over the eastern Pacific and the resultant increase in zonal SST contrast. In the CMIP ensemble, there is little change in the simulated zonal SST contrast in response to GHG and aerosol forcing; a relatively uniform SST warming

trend occurs across the entire tropics during 1977-2006 (cf Fig. 1, bottom). Another possible factor is that the interannual variability of SSTs during this 30-yr period including ENSO events, which are specified in AMIP but are unlikely to be synchronized with observations in CMIP, may also be contributing to their different rainfall trend amplitudes. In this regard, the CMIP ensemble possesses additional noise that is associated with each realization's natural coupled ocean-atmosphere variability, whereas only a single SST time history evolves in the AMIP ensemble. Finally, differences in intensity of rainfall trends could also reflect limitations in the AMIP experimental design (i.e, the specification of SSTs). Each of these hypotheses will be addressed further in Section 4 using additional suites of climate simulations.

b. SST and GHG signal-to-noise ratios for 30-yr change

Probability distribution functions (PDFs) of 30-yr trends are constructed using all individual AMIP and CMIP realizations in order to assess the roles of forcing and inherent noise in the 1977-2006 regional precipitation trends. The 30-yr change in mean-value of the PDF indicates the SST-forced and GHG-forced signals for AMIP and CMIP populations, respectively. The spread of the PDFs indicates the intensity of noise. Regarding the PDF spread, if our population samples had been drawn from a single model and if each realization of that model had been identically forced, the ensemble spread of 30-yr trends would estimate the intensity of internal atmospheric noise for AMIP, and of internal coupled ocean-atmospheric noise for CMIP. The PDFs, however, are not drawn from such a homogeneous population but from a multi-model dataset (in part to minimize biases in signal), and as such an additional factor influencing the spread is the model-dependent sensitivities to forcing. Also, individual models have not experienced identical forcing. For instance, CCM3 was forced with the Kaplan SSTs whereas

ECHAM4 was forced with HadISST, and different treatments are also used for anthropogenic aerosol forcing and natural forcings among CMIP models. Our estimates of the internal noise of 30-yr trends are therefore derived not only from the PDF spreads of the multi-model samples, but also from spreads computed using ensembles of individual models that were subjected to identical boundary conditions (see Table 1).

Precipitation trends averaged for five regions that experienced considerable change are highlighted in Fig. 3. For the Pacific equatorial cold tongue region (Fig. 3a), satellite estimates indicate a -28% decline in annual rainfall, which is qualitatively consistent in sign with the drying occurring in virtually all AMIP runs (blue curve). The mean AMIP decline of -15% is statistically significant at 95%, given the outwardly small atmospheric noise contribution to 30-yr trends. By comparison, the CMIP signal (red curve) is a +10% increase, and its population sample of change is mostly inconsistent with the strong observed decline. The observed decline is also greater than that occurring in any single AMIP realization, suggesting that the observed drying intensity resulted from a combination of SST forcing and strong atmospheric noise, though uncertainties in the satellite-derived trends and in the model sensitivities make more precise assessments difficult.

Coupled ocean-atmosphere noise is appreciably greater than atmospheric noise alone in contributing to variability of 30-year rainfall trends over the Pacific cold tongue region. This is readily apparent from a visual comparison of the AMIP and CMIP PDFs, and is further quantified in Table 1 where the spread of 30-yr trends in a multi-model average of unforced CMIP control runs is shown to be about four-fold greater than in a multi-model average of AMIP

runs. This indicates that the particular evolution of observed SSTs during 1977-2006 is materially important for understanding the observed rainfall trend, and that greater uncertainty in the CMIP simulations is almost certainly due to the effect of multiple SST trajectories and the rainfall sensitivity to them. How unusual the observed 1977-2006 SST trajectory in the tropical Pacific was, and the implications for both attribution and prediction are questions addressed further in subsequent sections.

Turning attention to the eastern hemisphere warm pool region, increased rainfall over the Indo-west Pacific occurs in response to both observed SST and GHG forcing during 1977-2006 (Fig. 3d). This being one of the globe's wettest areas, the 3% of climatological increase in the AMIP ensemble mean, though modest by comparison to other regional signals, nonetheless represents a substantial total change in rainfall. Further, the SST-induced wet signal is nearly triple that of the GHG-induced wet signal, which as mentioned earlier occurs despite the fact that the observed warm pool SST increase is of nearly identical strength to that occurring in the CMIP ensemble. It is plausible that the more substantial AMIP (and observed) rainfall increases could have resulted from a sensitivity to a particular interannual behavior of warm pool SSTs, though results in Section 4 will argue against such an effect for 1977-2006. Instead, it will be shown that there is strong warm pool rainfall sensitivity to the spatial pattern of SST change across the tropical ocean as a whole. Finally, regarding the PDFs, it is interesting to note the much smaller spread among individual CMIP simulated rainfall trends over the warm pool than that occurring over the Pacific cold tongue, and further that AMIP and CMIP spreads are nearly identical over the warm pool (see Table 1). There are two primary reasons to account for this. One is that the particular observed SST trajectory specified in AMIP over the Indo-west Pacific is close to the

simulated SST trajectory in CMIP occurring in response to external radiative forcing. Further, owing to weak unforced interannual-to-decadal warm pool SST variability compared to that occurring in the ENSO region, individual CMIP members experienced more consistent SST evolutions from run to run owing to the strength of their coherent externally forced warm pool warming.

Declines in annual precipitation over southwest North America and southeast Australia have raised concern among regional water resource managers that human-induced climate change is severely degrading their water supplies, above and beyond the hydrologic consequences also being exerted by warming surface temperatures. Our diagnosis provides little support to the speculation that these dryings have been of anthropogenic origins. The CMIP ensemble does indicate a modest -8% mean decline over southwest North America during 1977-2006 (Fig. 3b). However, the amplitude of that signal is only $\frac{1}{2}$ standard deviation of the noise in 30-yr precipitation trends associated with natural coupled variability, and there are nearly as many CMIP members yielding increased rainfall as decreased rainfall. It is therefore unlikely that the observed drying trend of -25% is due to human-induced emissions of GHGs and aerosols alone. In contrast, the AMIP ensemble signal of a -17% decline indicates that the observed drying was more consistent with variability of SSTs during the last 30-years. Importantly, 30 of the 36 AMIP simulations generate a drying trend. A relevant question is the extent to which the culpable SSTs are themselves of anthropogenic origin, and/or are attributes of natural ocean variability. Some insight comes from recalling that the observed minus CMIP SST trend pattern resembles La Niña (Fig. 1 bottom). Figure 4 shows the observed precipitation anomalies associated with La Niña, derived by regressing annual precipitation onto an annual index of cold

tongue SSTs. It is revealing that many aspects of the 30-yr observed precipitation trends over the Pacific sector and adjacent Americas are consistent with this regression pattern associated with a natural mode of the coupled climate system (cf. Fig. 2).

Neither SST nor anthropogenic forcing appear to have contributed to the observed decline in annual precipitation over southeast Australia (Fig. 3e). In fact, the CMIP ensemble mean yields a +10% increase in annual precipitation, and the vast majority of 21 individual models show a positive trend. It is also apparent from the prior regression analysis that a reduced southeast Australian precipitation is inconsistent with a cold state of the tropical east Pacific (see also Nicholls et al. 1996). Further, in so far as both AMIP and CMIP simulations experience a similar warming of Indo-west Pacific SSTs, the Australian drying is unlikely a consequence of warm pool warming either. The observed decrease of -11% could be reconciled, however, with a bout of strong atmospheric internal variability, which according to Table 1 would require a -1.4 standardized departure occurrence to match the amplitude of the observed 30-yr drying.

A final regional trend of interest is the outward recovery in Sahel rainfall (Fig. 3c). The +20% observed increase is close in intensity to the AMIP ensemble mean trend of +25% indicating that the recovery in Sahelian rainfall is consistent with the recent observed SST variability. This is not entirely surprising given the ability of AMIP experiments to replicate the region's rainfall trends for the 1950-1999 period (Giannini et al. 2003; Hoerling et al. 2006). The AMIP PDF is particularly skewed compared to that for other regions, the consequence of a single model whose Sahel rainfall exhibited only weak sensitivity to SSTs. Owing to this inter-model difference in signals, the atmospheric noise contribution to Sahel 30-yr trends is somewhat greater when

inferred from the PDF spread of the commingled 36 AMIP runs ($\sigma=22\%$) than from the spread averaged for each of the 4 models separately ($\sigma=16\%$, Table 1). Regarding the CMIP PDF, its ensemble mean shows only a slight Sahel rainfall increase, with nearly as many runs indicating a decrease as there are an increase. Further, the 3% ensemble mean increase pales in comparison to the estimated noise of 30-yr trends due to internal coupled ocean-atmosphere variability ($\sigma\sim 20\%$), implying that even the sign of the externally forced signal in Sahel rainfall change for 1977-2006 is uncertain.

4. The Role of SST Trends in Regional Precipitation Changes

The prior analysis indicates that a more realistic pattern and intensity of regional rainfall change is rendered when climate models are constrained by the trajectory of observed SST forcing than when they are constrained by the trajectory of external radiative forcing during 1977-2006. The implication is that the particular observed SST history was critical for the observed rainfall changes, and that the differences between AMIP and CMIP simulations might be explained by their different SST forcings.

We diagnose an additional suite of atmospheric model simulations to determine the role of the particular observed SST forcing for observed rainfall trends, and we test several hypotheses raised previously in section 3 regarding plausible explanations for differences between AMIP and CMIP. These model simulations were forced by the seasonal cycle of the observed and the CMIP ensemble SST changes for the 1977-2006 period. In one set, an SST anomaly equal to the linear trend of observed SSTs (Fig. 1, top) are specified, whereas the linear trend of the CMIP ensemble mean SSTs (Fig. 1, middle) is specified in a parallel set of the same models (see

section 2.2). Figure 5 shows the annual precipitation response calculated from the resulting 120-member multi-model ensemble.

One hypothesis raised in section 3 was that the pattern of regional rainfall trends reflected a sensitivity to *mean changes* in SSTs, and that the greater intensity of AMIP versus CMIP rainfall trends resulted principally from a sensitivity to differences in their mean patterns of SST change. The results on these additional experiments support a view that the patterns of mean SST changes were indeed of leading importance. Key features of simulated rainfall trends occurring in AMIP and CMIP in response to time-varying forcing during 1977-2006 can be recovered by subjecting the atmosphere to their 30-yr mean SST changes alone (compare Fig. 5 with the lower panels of Fig. 2). Also evident in Fig. 5 is the appreciable sensitivity of rainfall responses to the difference between observed and CMIP mean SST changes. For instance, a tropical Pacific dry response to the observed SST trend (Fig. 5, top) contrasts with a wet response to the CMIP SST trend (Fig. 5, bottom). Recall that eastern Pacific SSTs were observed to cool during 1977-2006 versus the CMIP ensemble mean SST warming (see Fig. 1), and the results of these additional experiments indicate that such local differences in SST change are the principal cause for differences between the AMIP and CMIP simulated 1977-2006 Pacific rainfall trends.

Even where the local SST trends are similar between observations and CMIP simulations, such as over the Indo-west Pacific warm pool, there are appreciable differences between the rainfall sensitivities to the two specified SST changes. Note in Fig. 5 the lack of any discernable warm pool rainfall response to the CMIP SST trend but a widespread increase in rainfall to the observed SST trend. This sensitivity mimics the AMIP versus CMIP differences of Fig. 2, and

suggests that the warm pool rainfall increase during 1977-2006 was linked to an *increased contrast in SSTs* between the east Pacific and warm pool region, and was not a response to local SST warming alone.

These additional experiments indicate that, at least for the 1977-2006 period, regional rainfall trends were especially sensitive to the pattern of 30-yr SST changes, and only secondarily sensitive to higher frequency components of the SST variations. This is not to say that the particular temporal behavior of observed SSTs wasn't important: for example, the observed slight cooling trend of the tropical east Pacific is in part a residual of higher frequency ENSO variability. It does indicate, however, the broad linearity of precipitation responses to SST forcing, and is consistent with similar findings in Schubert et al. (2004a) regarding their diagnosis of the role of SSTs in U.S. drought during the "Dust Bowl" decade of the 1930s.

A third issue raised in section 3 concerns the suitability of the AMIP approach, and to what extent the specification of SSTs in atmospheric GCMs may cause spurious sensitivity (e.g. Kumar and Hoerling 1997). It is almost certainly not coincidental that the AMIP ensemble mean rainfall trend resembles observed changes, indicating that diagnosis of the atmospheric response to SSTs is a powerful tool in understanding the more complicated coupled system. But a reasonable question is to what extent the feedbacks from coupled interactions are important for understanding the precipitation trends. While a thorough answer to this question is beyond the scope of our study, some insight comes from a comparison of Figs. 2 and 5. Many features of the ensemble-mean precipitation trends simulated in the coupled CMIP models in response to GHG forcing are reproduced by subjecting uncoupled atmospheric models to the linear trend of

the CMIP SSTs. We further note that the amplitude of rainfall responses to the specified CMIP SSTs is subdued, and is consistent with the weak amplitude occurring in the fully coupled models. This is not to discount the role of other effects that anthropogenic climate forcings (i.e. aerosols via its direct and indirect effects) included in the CMIP simulations can exert on regional precipitation changes (e.g., Ming and Ramaswamy, 2009). We are, however, reasonably assured that the AMIP ensemble precipitation trends offer a realistic diagnosis of the role of SSTs in the fully coupled observed system during 1977-2006, and that they also provide insight on the causes for rainfall trends in the CMIP simulations.

Further evidence for the importance of the SST trend is provided in Figure 6, which compares the differences in responses to the 1977-2006 observed trend versus the CMIP ensemble SST trends (left) to the maps of the trend differences between AMIP and CMIP ensemble means (right). In addition to the precipitation (top), also shown is the difference in 200 hPa height responses (bottom) which provides a dynamical framework for interpreting some of the regional precipitation differences. The outstanding feature is a wavetrain of alternating low and high pressure that arches across the North and South Pacific Oceans that is symptomatic of the teleconnections associated with cold sea surface temperature forcing and reduced rainfall in the tropical east Pacific (cf. Fig. 4). The drier conditions over the subtropical North American and South American continent are likewise consistent with storm track shifts typically associated with such tropically forced teleconnections (e.g., Trenberth et al. 1998). Likewise, we suspect that a broad area of positive height differences extending from eastern Canada, the North Atlantic and northern Europe is associated with the greater observed warming trend of North Atlantic SSTs compared to CMIP (see Fig. 1). Indeed, a regression of annual precipitation and

200 hPa heights onto an index of North Atlantic SSTs reveals a prominent NAO-like pattern together with an increase in Sahel rainfall (not shown). The comparison between the left and right hand maps of Fig. 6 confirms that the differences between AMIP and CMIP simulations can be explained as resulting mostly from the atmospheric sensitivity to their different 30-yr SST trends. The principal features distinguishing the AMIP from the CMIP simulated trends have their counterparts in the differences between atmospheric responses to their respective SST trends alone. For precipitation (200hPa heights) the two maps are correlated at 0.68 (0.59), confirming that much of the difference between AMIP and CMIP simulated rainfall and circulation trends is attributable to a sensitivity to differences in their respective SST trends alone.

5. Summary and Discussion

The nature and causes for observed regional precipitation trends during 1977-2006 have been diagnosed. We focused on a decrease in annual rainfall over the equatorial east Pacific Ocean, the subtropical north and south Pacific and adjacent western portions of the Americas. We also assessed the causes for eastern hemisphere rainfall changes including increases over the Sahel, and much of the tropical Indo-Pacific warm pool, and a notable drying over southeast Australia. Two ensemble suites of climate models were analyzed in order to determine the effect of known forcings on these regional trends in particular; the influence of external radiative forcing was diagnosed from the coupled models (CMIP3) that were done in support of the Intergovernmental Panel on Climate Change (IPCC) Fourth Assessment Report (IPCC 2007), and the influence of observed SST forcing was estimated from atmospheric models subjected to the monthly variations of global SSTs but using climatological fixed GHG and aerosol concentrations.

Forcing by observed sea surface temperature variability was a major factor explaining the pattern of regional precipitation trends during 1977-2006, especially over the tropics. The western hemisphere pattern dominated by drying was consistent with an atmospheric response to modest cooling of equatorial east Pacific SSTs, a sensitivity resembling the atmospheric response observed on interannual time scales in association with La Niña. The eastern hemisphere pattern of increased warm pool precipitation was consistent with a warming of local SSTs, and was further consistent with a sensitivity to an intensified zonal contrast of tropical SSTs during 1977-2006. Regarding recovery of Sahel rainfall, we argued that a strong warming of North Atlantic SSTs during recent decades was a leading factor. It induced a Sahelian response that is consistent with the predilection of the Atlantic ITCZ to intensify and shift poleward in concert with a warming North Atlantic (e.g. Folland et al., 1986). We found little sensitivity of Australian rainfall trends to SST forcing in our suite of AMIP models, leading to the conclusion that the southeast Australian drying in recent decades was mostly inconsistent with SST forcing.

The SST forced rainfall changes, diagnosed from AMIP simulations, was generally not symptomatic of an effect attributable to human-induced emissions of greenhouse gases and aerosols. The simulated rainfall trends in CMIP simulations were found to be considerably weaker than those observed and also weaker than occurring in the AMIP simulations. Further, the spatial pattern of externally forced rainfall changes differed appreciably from observations. For instance, CMIP simulated increases in rainfall over the tropical Pacific and southeastern Australia were contrary to observed drying in these areas. Even over the Indo-Pacific warm pool where observed and CMIP simulated SST warmings were very similar, CMIP rainfall increases

were much weaker. Additional model experiments clarify that the discrepancy between observed and externally forced rainfall changes during 1977-2006 resulted mostly from the differences between their patterns of 30-yr SST change, and the atmospheric sensitivity to them. Most notably, a warming of the tropical east Pacific in response to GHG and aerosol forcing contributed to locally increased rainfall and related remote impacts in CMIP that were contrary to the effect which observed east Pacific SST cooling exerted. Overall, the externally forced pattern of uniform SST warming throughout the tropical and subtropical latitudes was found to be ineffective in inducing regional rainfall responses with appreciable amplitude. By contrast, the observed pattern of SST change that was typified by increases in gradients was a more effective driver of regional rainfall trends.

How then are the observed changes in SSTs over the recent three decades to be understood, and how are they to be reconciled with the ocean's response to GHG and aerosol forcing? Two features of the 1977-2006 SST trend that impacted tropical rainfall trends especially were the equatorial east Pacific cooling and the warming of the Indo-west Pacific. Shown in Fig. 7 are estimates of statistical probabilities of 30-yr SST trends over these two areas, on the one hand due to the trajectory of external radiative forcing during 1977-2006 (solid curves) and on the other hand due to internal coupled ocean-atmosphere noise alone (dashed curves). The solid curves are estimates of the PDFs based on the 21 forced CMIP simulations, whereas the dashed curves are based on 30-yr trends calculated from unforced, pre-industrial control simulations of 4 CMIP models having realistic ENSO variability (see section 2). The observed warm pool warming of $+0.4^{\circ}\text{C}$ is equal to the mean CMIP simulated warming intensity, and significantly exceeds the spread of 30-yr trends occurring in both forced and unforced coupled models

($\sigma \sim 0.1^\circ\text{C}$). Thus, a change in warm pool SSTs has likely been detected during the prior 30 years, one that is largely attributable to external radiative forcing (see also Knutson et al. 2006, Hoerling et al. 2004). In contrast, the observed cold tongue SSTs cooled by -0.1°C , which was a considerable departure from the $+0.4^\circ\text{C}$ mean CMIP simulated warming. No single CMIP run generated a 30-yr cooling for 1977-2006. The large spread of 30-yr trends in the cold tongue region ($\sigma \sim 0.2^\circ\text{C}$) could indicate though that the observed cooling resulted from a particularly strong occurrence of natural variability.

One of the important scientific issues is how the tropical oceans will respond to anthropogenic forcing, and in particular whether the spatial pattern of change will lead to either an increased or decreased zonal contrast of SSTs between the warm pool and cold tongue regions (e.g., Vecchi et al. 2008). Clearly, the time-mean zonal SST gradient over the tropical Pacific has increased in the past 30 years, and this particular trajectory appears to be a low probability state of the CMIP ensemble simulations. The extent to which this reflects a bias in how coupled models respond to anthropogenic forcing, whether the CMIP models underestimate the intensity of internal coupled ocean-atmosphere variability, or whether an unusual occurrence of natural variability has simply masked the anthropogenic signal are open questions. The veracity of a particular interpretation has important implications however, since our results have shown that the particular trajectory of SSTs during 1977-2006 materially influenced the patterns of regional rainfall change, and that significantly different regional rainfall patterns would have evolved had nature instead adopted the path of the CMIP ensemble SST change. These issues pose obvious challenges for predictions of regional rainfall change for coming decades. It is likely that mean changes in SSTs will continue to strongly influence the pattern of precipitation changes at regional scales,

for instance as shown in a recent study of the joint evolution of sea surface temperature and rainfall within coupled model projections using 21st Century GHG emission scenarios (Xie et al. 2009). The challenge is predicting the SST trajectory itself, and the nascent efforts of initialized decadal forecasts (e.g., Smith et al. 2007; Keenlyside et al. 2008) will be seeking to capitalize on knowledge of the initial state of the ocean in order to render forecasts for future decades that might improve on the uninitialized projections of IPCC-type models alone.

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Table 1. The 1977-2006 observed and simulated regional precipitation trends, expressed as the % change relative to observed and modeled annual climatological means. The changes in means are based on the 4 model average for AMIP and the 21 model average for CMIP. The standard deviations, or noise, are based on two estimates. The first number is the spread among the 36 AMIP realizations and the 21 CMIP simulations. The number in parenthesis for AMIP is the spread among the 9-members for each model, and then averaged across the four models. The number in parenthesis for CMIP is the spread among 10 samples of 30-yr trends computed for control integrations of 4 CMIP models, and then averaged across the four models. The observed trends over the oceanic regions that comprise the cold tongue and warm pool are for 1979-2006. Figure 4 shows the geographical location of each region.

	ΔMean (%Climo)			Std Dev (%Climo)	
	OBS	AMIP	CMIP	AMIP	CMIP
Cold Tongue	-28	-15	+10	6 (4)	16 (27)
Warm Pool	+8	+3	+1	2 (2)	2 (2)
Sahel	+20	+25	+3	22 (16)	20 (23)
SW U.S.	-25	-17	-8	21 (20)	17 (20)
SE Australia	-11	-1	+10	8 (8)	13 (12)

Figure Captions:

Figure 1. The observed (top), CMIP ensemble mean simulated (middle), trends in annual sea surface temperatures during 1977-2006, and their difference (bottom). The top panels are spatially correlated at 0.32 over the world oceans.

Figure 2. The observed (top), AMIP ensemble mean simulated (middle), and CMIP ensemble mean simulated (bottom) 30-year trends in annual precipitation during 1977-2006. The observed and AMIP (CMIP) rainfall trend patterns are correlated at 0.32 (0.19) over the globe, whereas AMIP and CMIP patterns have near-zero correlation.

Figure 3. The probability distribution functions of the 1977-2006 precipitation trends occurring in the individual 36 AMIP simulations (blue PDFs), and the 21 CMIP simulations (red curve). Five regions are highlighted that experienced appreciable change in their observed precipitation during 1977-2006: (a) the equatorial Pacific cold tongue, southwest North America (b), the Sahel (c), the Indo-Pacific warm pool (d), and southeast Australia (e). Gray bands denote the geographical areas over which 30-yr precipitation trends were computed. Large gray tick on each PDF denotes the observed 30-yr change, whereas the 30-yr change for each individual AMIP (CMIP) member is denoted by blue (red) ticks.

Figure 4. The observed linear regression of annual precipitation (color shades) and 200 hPa heights (contours) upon an index of equatorial east Pacific SSTs for 1977-2006. The index is of the cold tongue region from 5°N-5°S, 180°-South America. Regression shown is for a -1

standardized departure of the SST index corresponding to La Niña conditions. Precipitation is % of annual climatology.

Figure 5. The annual precipitation response to specified SST anomalies corresponding to the 1977-2006 observed SST change (top), and the 1977-2006 CMIP ensemble mean SST change. Based on a 120-member, 4-model ensemble average. The anomalies are % of annual model climatology.

Figure 6. The difference in the responses to specified observed versus CMIP 1977-2006 SST changes (left). The difference between the AMIP ensemble mean simulated and CMIP ensemble mean simulated 30-year trends during 1977-2006. Precipitation (200 hPa height) differences shown on top (bottom) panels. The left panels are based on a 120-member, 4-model ensemble average. The right panels are based on the 36-member AMIP and 21-member CMIP simulations for 1977-2006. The precipitation anomalies are % of annual model climatology. Spatial correlation between top (bottom) two panels is 0.68 (0.59).

Figure 7. The probability distribution functions of the 1977-2006 sea surface temperature trends occurring in the 21 CMIP simulations. Two regions are shown: the Indo-Pacific warm pool (top, red curve), and the equatorial Pacific cold tongue (bottom, blue curve). The geographical areas over which 30-yr precipitation trends were computed are identical to those in Fig. 4. The probability distribution functions of 30-yr SST trends derived from pre-industrial control integrations of 4 CMIP models is superimposed in the dashed green curves. A large tick mark on

each PDF denotes the observed 30-yr change, whereas the 30-yr change for each individual CMIP member is denoted by small tick marks.

Annual Sea Surface Temperature Trend

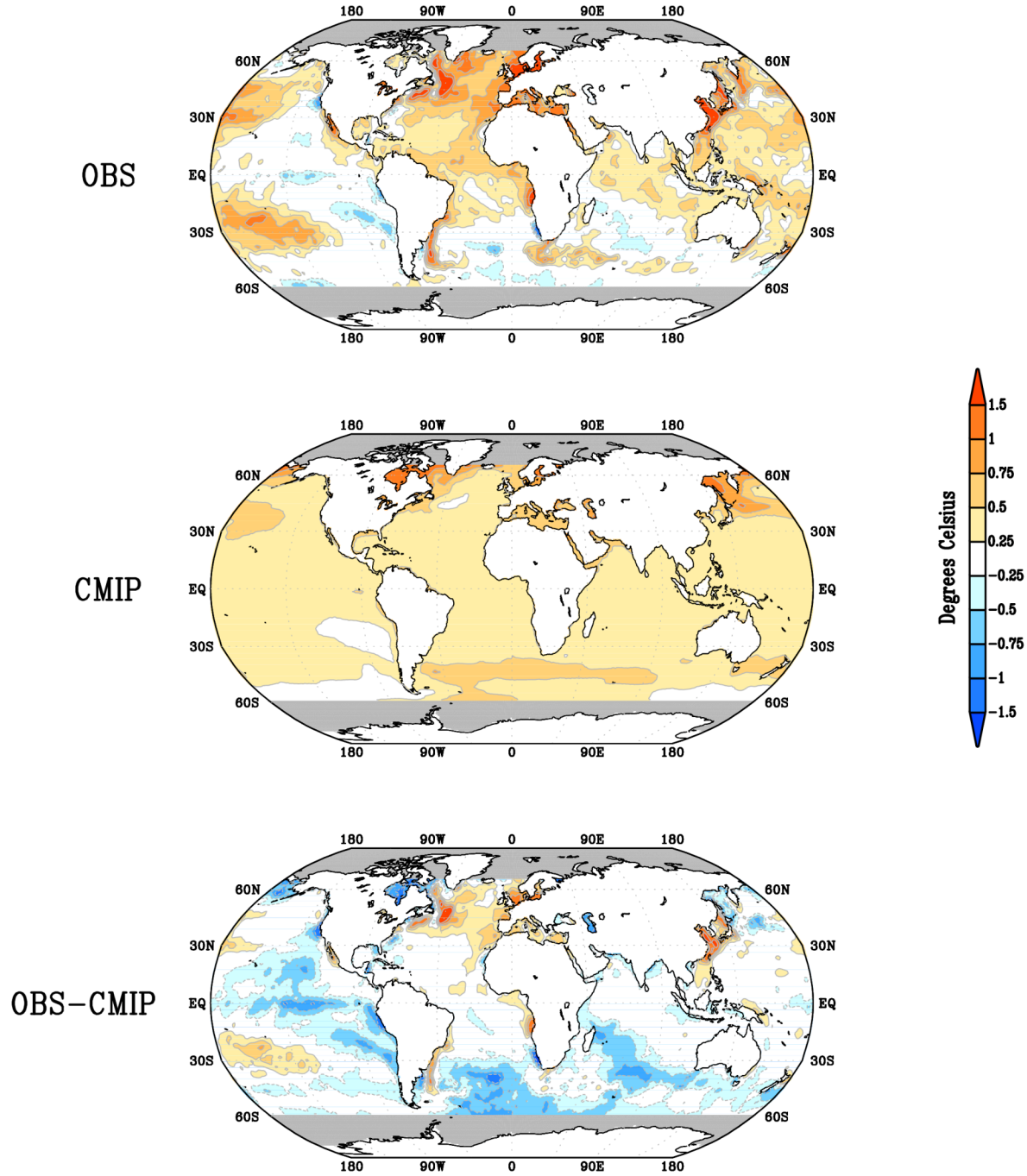


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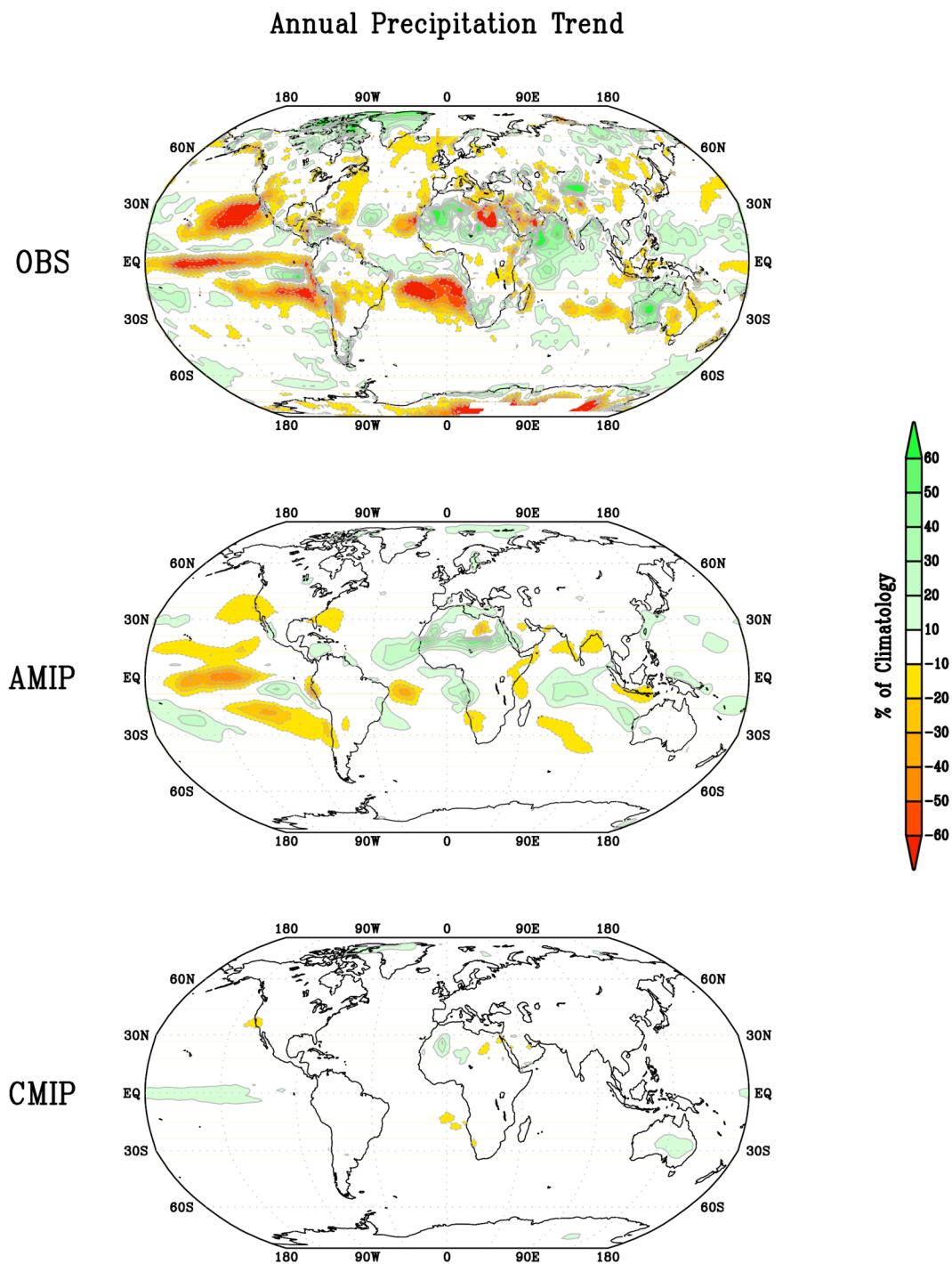


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Probability Distribution of Simulated Precipitation Trend 1977–2006

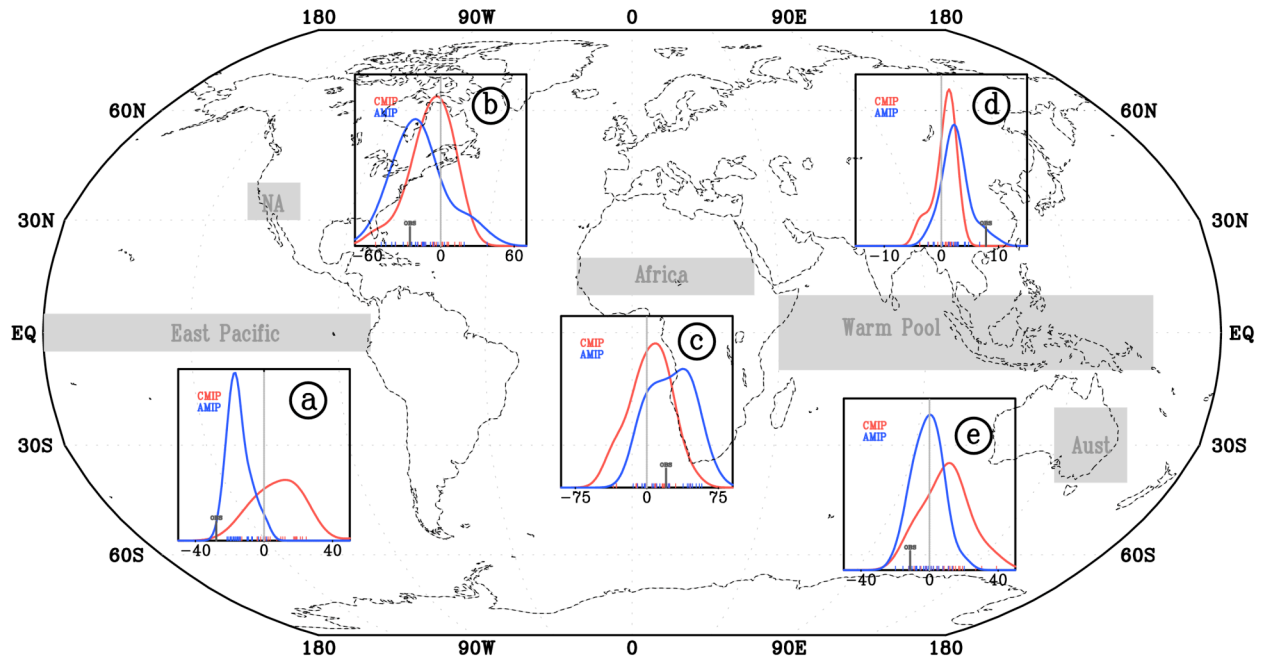


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Cold Tongue Precipitation/200hPa Height Regressions

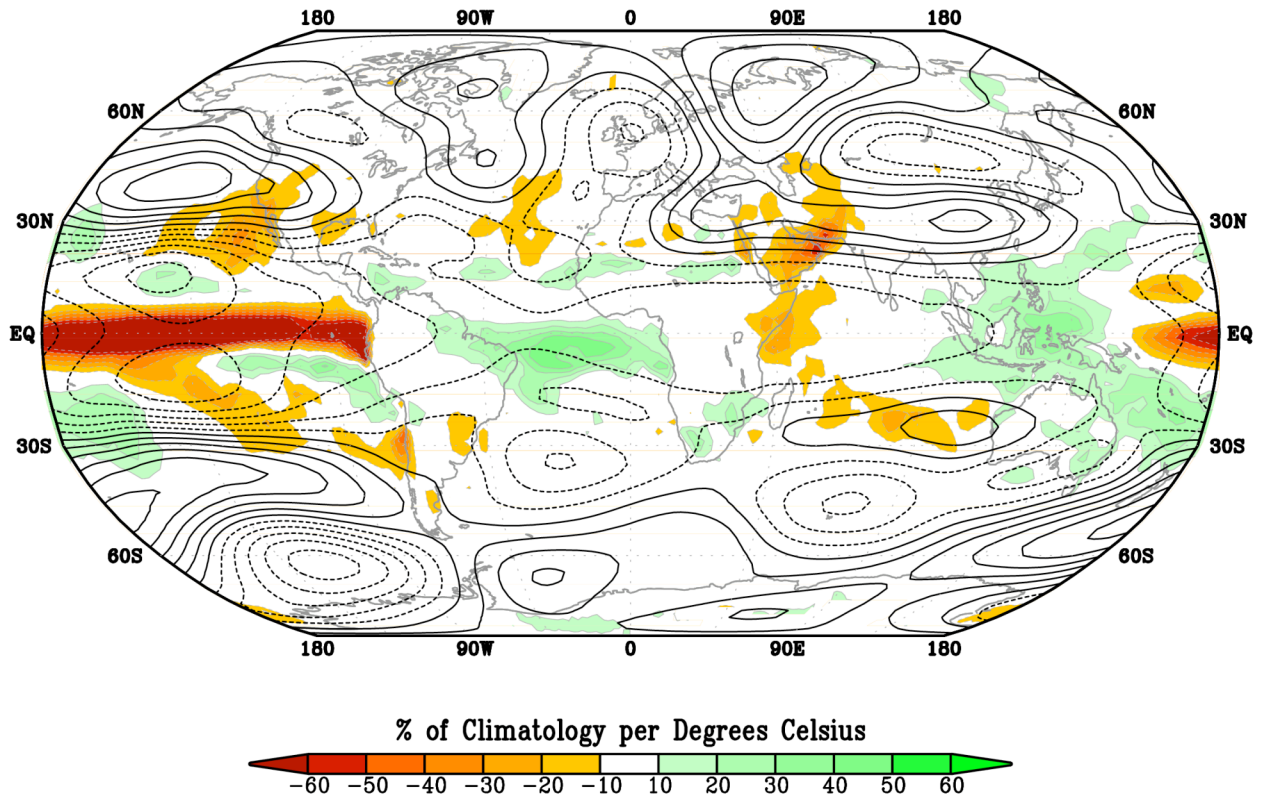


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Precipitation Response to SST Trends

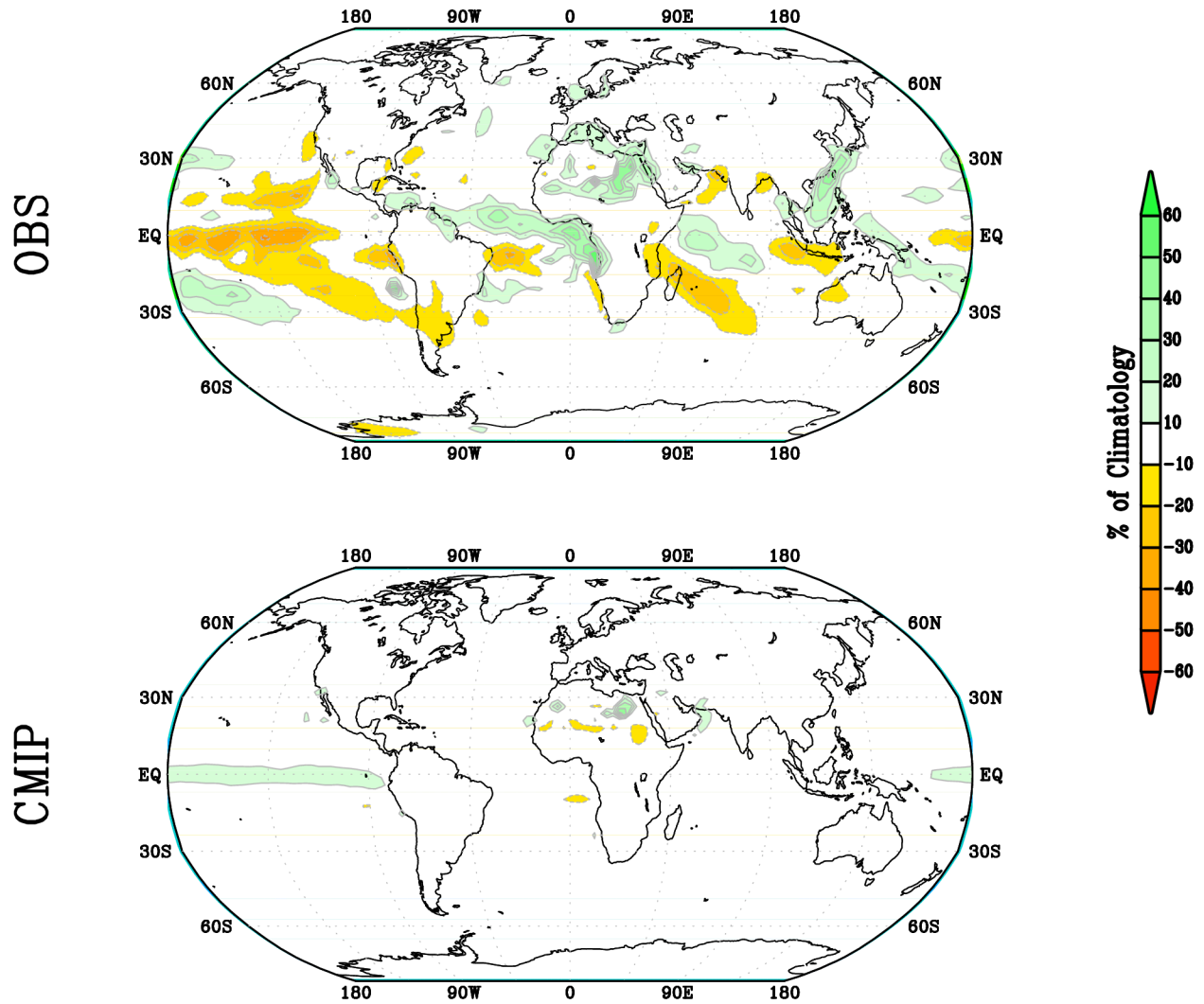


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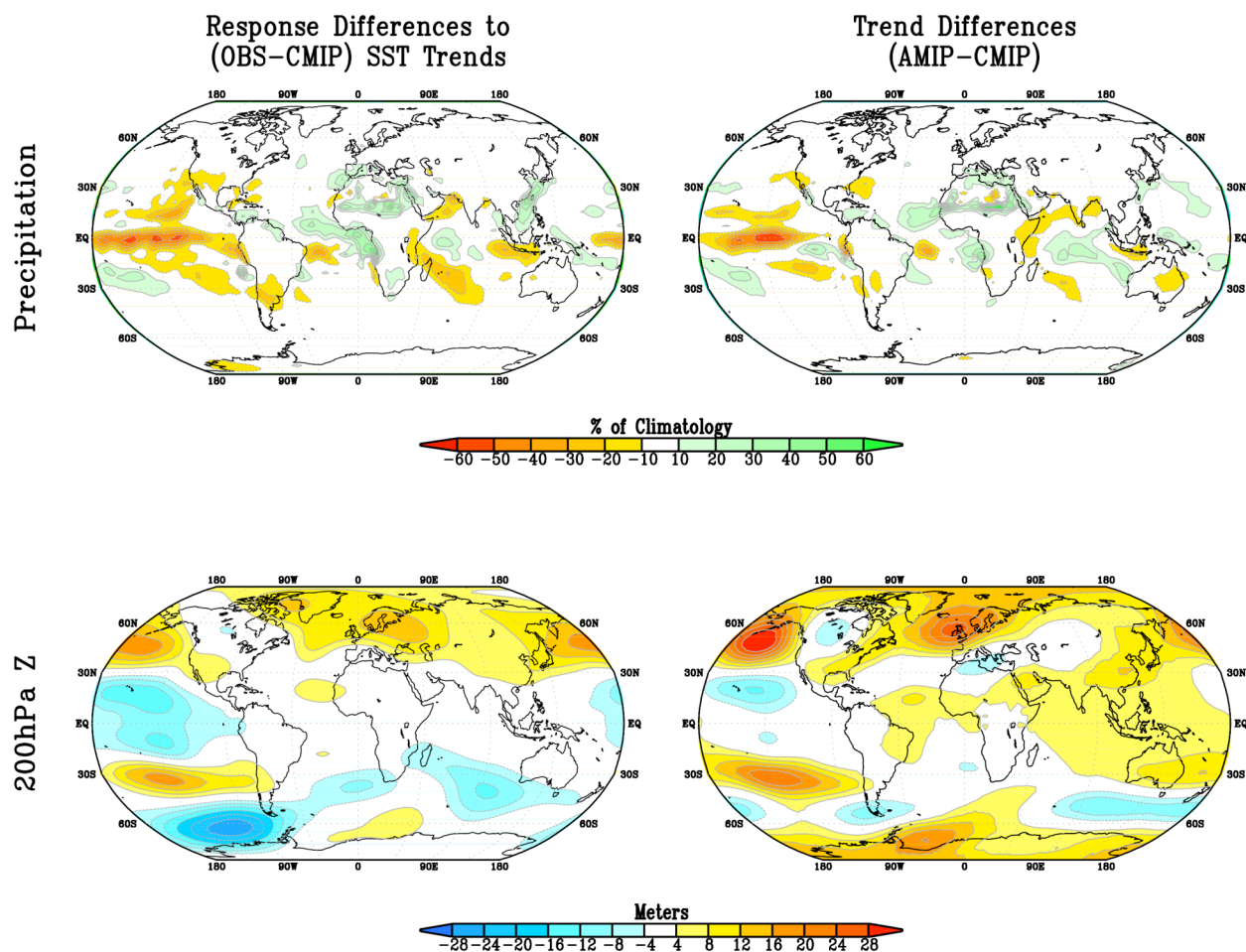


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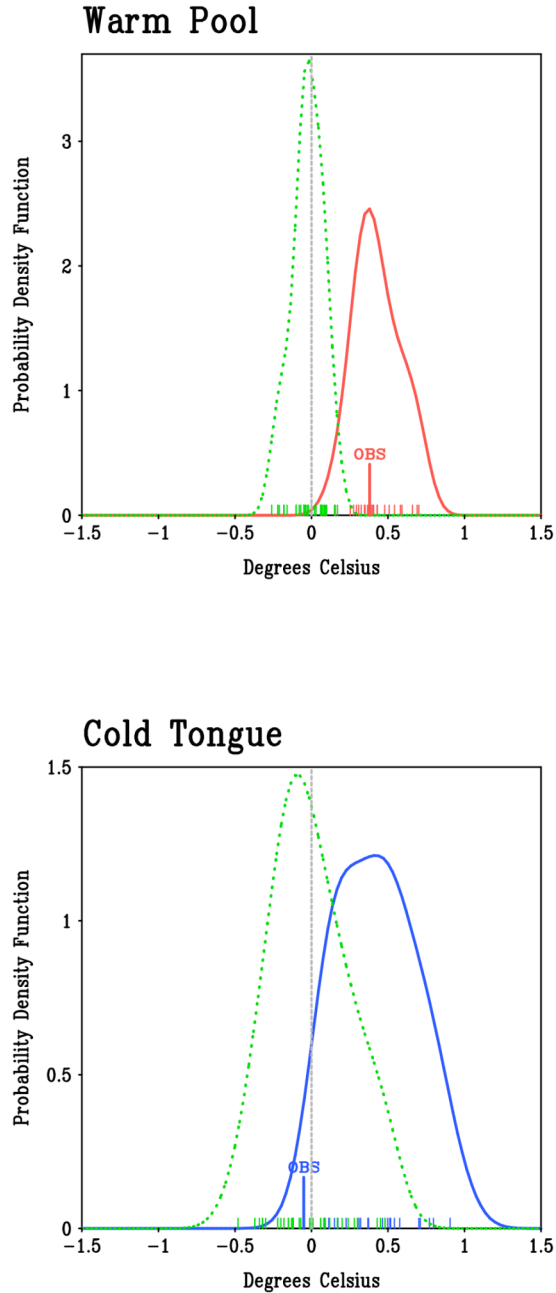


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